## Floating Point

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## Fractional binary numbers

■ What is $\mathbf{1 0 1 1 . 1 0 1}_{2}$ ?

## Fractional Binary Numbers: Examples

```
| Value Representation
    5 3/4 101.112
    27/8 10.1112
    17/16 1.01112
```

- Observations
- Divide by 2 by shifting right (unsigned)
- Multiply by 2 by shifting left
- Numbers of form $0.111111 \ldots 2$ are just below 1.0
- $1 / 2+1 / 4+1 / 8+\ldots+1 / 2^{i}+\ldots \rightarrow 1.0$


## Today: Floating Point

■ Background: Fractional binary numbers

- IEEE floating point standard: Definition
- Example and properties
- Rounding, addition, multiplication
- Floating point in C
- Summary

Fractional Binary Numbers


- Bits to right of "binary point" represent fractional powers of 2
- Represents rational number:

$$
\sum_{k=-1}^{1} b_{k} \times 2^{k}
$$

## Representable Numbers

## - Limitation \#1

- Can only exactly represent numbers of the form $x / 2^{k}$ - Other rational numbers have repeating bit representations
- Value Representation
- 1/3 0.0101010101 [01] ... 2
- $1 / 5 \quad 0.001100110011[0011]$... 2
- 1/10 0.0001100110011[0011] ... 2
- What if the number of bits is limited?
- "Fixed point": just one setting of binary point within the $w$ bits
- Limited range of numbers (bad for very small or very large values)


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## IEEE Floating Point

- IEEE Standard 754
- Established in 1985 as uniform standard for floating point arithmetic
- Before that, many idiosyncratic formats
- Supported by all major CPUs


## - Driven by numerical concerns

- Nice standards for rounding, overflow, underflow
- A lot of work to make fast in hardware
- Numerical analysts predominated over hardware designers in defining standard


## Floating Point Representation

## - Numerical Form:

## $(-1)^{5} M 2^{E}$

- Sign bit $s$ determines whether number is negative or positive
- Significand $M$ normally a fractional value in range [1.0,2.0).
- Exponent $E$ weights value by power of two


## - Encoding

- MSB $s$ is sign bit $s$
- $\exp$ field encodes $E$ (but is not equal to $E$ )
- frac field encodes $M$ (but is not equal to $M$ )


Bryant and O'Hallaronn, Computer Systems: A Programmer's Perspective, Third Edition $^{\prime}$
"Normalized" (Normal) Values $v=(-1)^{s} M 2^{E}$

- When: $\exp \neq 000$... 0 and $\exp \neq 111$... 1
- Exponent coded as a biased value: $E=$ Exp - Bias
- Exp: unsigned value of exp field
- Bias $=2^{k-1}-1$, where $k$ is number of exponent bits
- Single precision: 127 (Exp: 1...254, E: -126...127)
- Double precision: 1023 (Exp: 1...2046, E: -1022...1023)

■ Significand coded with implied leading 1: $M=1 . x x x \ldots . . x_{2}$

- xxx...x: bits of frac field
- Minimum when frac=000... 0 ( $\mathrm{M}=1.0$ )
- Maximum when frac=111... 1 ( $\mathrm{M}=2.0-\varepsilon$ )
- Get extra leading bit for "free"


## Precision options

- Single precision: 32 bits

| s | $\exp$ | frac |  |
| :--- | :--- | :--- | :--- |
| 1 | 8-bits | 23-bits |  |

- Double precision: 64 bits

- Extended precision: $\mathbf{8 0}$ bits (older Intel only)

smant and O'Hallaron, Computer Systems: A Programmer's Perspective, Third Edtion


## Denormalized Values

## $\mathrm{V}=(-1)^{\mathrm{S}} \boldsymbol{M} 2^{E}$

$E=1$ - Bias

- Condition: $\exp =000 . . .0$
- Exponent value: $E=1$ - Bias (instead of $E=0-$ Bias)
- Significand coded with implied leading $0: M=0 . x x x$...x
- xxx x: bits of frac
- Cases
- $\exp =000 \ldots 0$, frac $=000 \ldots$
- Represents zero value
- Note distinct values: +0 and -0 (why?)
- exp $=000 \ldots 0$, frac $=000 . . .0$
- Numbers closest to 0.0
- Equispaced


## Visualization: Floating Point Encodings



## Tiny Floating Point Example

| $s$ | $\exp$ | frac |
| :---: | :---: | :---: |
| 1 | 4-bits | 3-bits |

## - 8-bit Floating Point Representation

- the sign bit is in the most significant bit
- the next four bits are the exponent, with a bias of 7
- the last three bits are the frac


## ■ Same general form as IEEE Format

- normalized, denormalized
- representation of $0, \mathrm{NaN}$, infinity


## Special Values

■ Condition: $\exp =111 . . .1$

- Case: $\exp =111 \ldots 1$, frac $=000 \ldots 0$
- Represents value $\infty$ (infinity)
- Operation that overflows

Both positive and negative

- E.g., $1.0 / 0.0=-1.0 /-0.0=+\infty, 1.0 /-0.0=-\infty$
- Case: $\exp =111$...1, frac $\neq 000$... 0
- Not-a-Number (NaN)
- Represents case when no numeric value can be determined
- E.g., sqrt( -1 ), $\infty-\infty, \infty \times 0$


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| Dynamic Range (Positive Only) |  |  |  |  |  |  | $\begin{gathered} \mathrm{V}=(-1)^{\mathrm{S}} M 2^{E} \\ \mathrm{n}: E=\operatorname{Exp}-\text { Bias } \\ d: E=1-\text { Bias } \end{gathered}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | exp f | frac <br> 000 |  | value |  |  |
|  |  | 0000 | 001 | -6 | 1/8*1/64 | $=1 / 512$ | closest to zero |
| Denormalized numbers |  | 0000 | 010 | -6 | 2/8*1/64 | $=2 / 512$ |  |
|  |  | 0000 | 110 | -6 | 6/8*1/64 | $=6 / 512$ |  |
|  |  | 0000 | 111 | -6 | 7/8*1/64 | $=7 / 512$ |  |
| Normalized numbers | 00 | 0001 | 000 | -6 | 8/8*1/64 | $=8 / 512$ | smallest norm |
|  |  | 0001 | 001 | -6 | 9/8*1/64 | $=9 / 512$ |  |
|  |  | 0110 | 110 | -1 | 14/8*1/2 | = $14 / 16$ |  |
|  |  | 0110 | 111 | -1 | 15/8*1/2 | = $15 / 16$ | closest to 1 below |
|  |  | 0111 | 000 | 0 | 8/8*1 | = 1 |  |
|  |  | 0111 | 001 | 0 | 9/8*1 | = 9/8 | closest to 1 above |
|  |  | 0111 | 010 | 0 | 10/8*1 | $=10 / 8$ | closest to 1 above |
|  |  | 1110 | 110 | 7 | 14/8*128 | $=224$ |  |
|  |  | 1110 | 111 | 7 | 15/8*128 | $=240$ | largest norm |
|  |  | 1111 | 000 | n/a | inf |  |  |

## Distribution of Values

- 6-bit IEEE-like format
- e = 3 exponent bits
- $f=2$ fraction bits
- Bias is $2^{3-1}-1=3$

- Notice how the distribution gets denser toward zero.



## Special Properties of the IEEE Encoding

- FP Zero Same as Integer Zero
- All bits = 0
- Can (Almost) Use Unsigned Integer Comparison
- Must first compare sign bits
- Must consider - $0=0$
- NaNs problematic
- Will be greater than any other values
- What should comparison yield?
- Otherwise OK
- Denorm vs. normalized
- Normalized vs. infinity

Floating Point Operations: Basic Idea

- $x+y=\operatorname{Round}(x+y)$
- $\mathbf{x} \times_{f} \mathbf{y}=$ Round $(\mathbf{x} \times \mathrm{y})$


## - Basic idea

- First compute exact result
- Make it fit into desired precision
- Possibly overflow if exponent too large
- Possibly round to fit into frac


## Distribution of Values (close-up view)

- 6-bit IEEE-like format
- $e=3$ exponent bits
- $f=2$ fraction bits
- Bias is 3



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## Rounding

- Rounding Modes (illustrate with \$ integer rounding)

|  | $\mathbf{\$ 1 . 4 0}$ | $\mathbf{\$ 1 . 6 0}$ | $\mathbf{\$ 1 . 5 0}$ | $\mathbf{\$ 2 . 5 0}$ | $\mathbf{- \$ 1 . 5 0}$ |
| :--- | :--- | :--- | :--- | :--- | :--- |
| - Towards zero | $\$ 1$ | $\$ 1$ | $\$ 1$ | $\$ 2$ | $-\$ 1$ |
| - Round down $(-\infty)$ | $\$ 1$ | $\$ 1$ | $\$ 1$ | $\$ 2$ | $-\$ 2$ |
| - Round up $(+\infty)$ | $\$ 2$ | $\$ 2$ | $\$ 2$ | $\$ 3$ | $-\$ 1$ |
| - Nearest Even (default) | $\$ 1$ | $\$ 2$ | $\$ 2$ | $\$ 2$ | $-\$ 2$ |

- What are the different modes good for?
- Towards zero: compatible with C integer behavior
- Round down/up: maintain conservative intervals
- Nearest even: unbiased, minimal error


## Closer Look at Round-To-Even <br> - Default Rounding Mode

- Hard to get any other kind without dropping into assembly
- All others are statistically biased
- Sum of set of positive numbers will consistently be over- or underestimated


## - Applying to Other Decimal Places / Bit Positions

- When exactly halfway between two possible values
- Round so that least significant digit is even
- E.g., round to nearest hundredth

| 7.8949999 | 7.89 | (Less than half way) |
| :--- | :--- | :--- |
| 7.8950001 | 7.90 | (Greater than half way) |
| 7.8950000 | 7.90 | (Half way-round up) |

$7.8850000 \quad 7.88 \quad$ (Half way-round down)

## Exercise break: FP and money?

- Your sandwich shop uses single-precision floating point for sales amounts
■ Need to apply a Minneapolis sales tax of $7.75 \%$, rounded up to the nearest cent
- On $\$ 4.00$ purchase, compute:
- round_up $(4.00 * 0.0775 * 100)=32$ cents
- Correct tax is 31 cents
- What went wrong?
- Note: $0.0775=31 / 400$ exactly


## FP Multiplication

- $(-1)^{s 1} M 12^{E 1} \times(-1)^{s 2}$ M2 $2^{E 2}$
- Exact Result: $(-1)^{\mathrm{s}} \boldsymbol{M} \mathbf{2}^{\mathrm{E}}$
- Sign s: $\quad$ 1 ^s2
- Significand $M$ : $M 1 \times M 2$
- Exponent $E: \quad E 1+E 2$
- Fixing
- If $M \geq 2$, shift $M$ right, increment $E$
- If $E$ out of range, overflow
- Round $M$ to fit frac precision
- Implementation
- Biggest chore is multiplying significands

Floating Point Addition


## Mathematical Properties of FP Add

## - Compare to those of Abelian Group

- Closed under addition? Ye
- But may generate infinity or NaN
- Commutative? Yes
- Associative? No
- Overflow and inexactness of rounding
- $(3.14+1 \mathrm{e} 10)-1 \mathrm{e} 10=0,3.14+(1 \mathrm{e} 10-1 \mathrm{e} 10)=3.14$
- 0 is additive identity?
- Every element has additive inverse? Yes
- Yes, except for infinities \& NaNs Almost
- Monotonicity
- $\mathrm{a} \geq \mathrm{b} \Rightarrow \mathrm{a}+\mathrm{c} \geq \mathrm{b}+\mathrm{c}$ ? Almost
- Except for infinities \& NaNs


## Mathematical Properties of FP Mult

- Compare to Commutative Ring
- Closed under multiplication? Yes
- But may generate infinity or NaN
- Multiplication Commutative? Yes
- Multiplication is Associative? No
- Possibility of overflow, inexactness of rounding
- Ex: $(1 \mathrm{e} 20 * 1 \mathrm{e} 20) * 1 \mathrm{e}-20=\mathrm{inf}, 1 \mathrm{e} 20 *(1 \mathrm{e} 20 * 1 \mathrm{e}-20)=1 \mathrm{e} 20$
- 1 is multiplicative identity?

Yes

- Multiplication distributes over addition?

No

- Possibility of overflow, inexactness of rounding
- $1 \mathrm{e} 20 *(1 \mathrm{e} 20-1 \mathrm{e} 20)=0.0,1 \mathrm{e} 20 * 1 \mathrm{e} 20-1 \mathrm{e} 20 * 1 \mathrm{e} 20=\mathrm{NaN}$
- Monotonicity
- $a \geq b \quad \& c \geq 0 \Rightarrow a^{*} c \geq b^{*} c$ ?

Almost

- Except for infinities \& NaNs


## Floating Point in C

- C Guarantees Two Levels
-float single precision
-double double precision
■ Conversions/Casting
- Casting between int, float, and double changes bit representation
- double/float $\rightarrow$ int
- Truncates fractional part
- Like rounding toward zero
- Not defined when out of range or NaN: Generally sets to TMin
- int $\rightarrow$ double
- Exact conversion, as long as int has $\leq 53$ bit word size
- int $\rightarrow$ float
- Will round according to rounding mode
$\qquad$


## Summary

■ IEEE Floating Point has clear mathematical properties

- Represents numbers of form $\mathbf{M} \times \mathbf{2}^{\mathrm{E}}$
- One can reason about operations independent of implementation
- As if computed with perfect precision and then rounded
- Not the same as real arithmetic
- Violates associativity/distributivity
- Makes life difficult for compilers \& serious numerical applications programmers


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## Floating Point Puzzles (full)

## - For each of the following C expressions, either:

- Argue that it is true for all argument values
- Explain why not true
- $x=($ int $)(f l o a t) \quad x$
- $x==$ (int) (double) $x$
int $x=\ldots ; \quad \cdot \mathbf{f}=$ (float) (double) $f$
float $f=\ldots$; $\quad$ d == (double) (float) $d$
double $d=\ldots$;
- $\mathbf{f}==-(-f)$;
- $2 / 3==2 / 3.0$

Assume neither $\quad d<0.0 \Rightarrow((d * 2)<0.0)$
$d$ nor $f$ is $\mathrm{NaN} \quad \Rightarrow \mathrm{d}>\mathrm{f} \quad \Rightarrow-f>-\mathrm{d}$

- $d * d>=0.0$
- $(d+f)-d==f$


## Additional Slides

## Creating Floating Point Number

- Steps
- Normalize to have leading 1
- Round to fit within fraction
- Postnormalize to deal with effects of rounding


## - Case Study

- Convert 8-bit unsigned numbers to tiny floating point format

Example Numbers
$128 \quad 10000000$
00001101
00010001
00010011
10001010 00111111

## Normalize

## - Requirement

- Set binary point so that numbers of form 1.xxxxx
- Adjust all to have leading one
- Decrement exponent as shift left

| Value | Binary | Fraction | Exponent |
| ---: | :--- | :--- | :--- |
| 128 | 10000000 | 1.0000000 | 7 |
| 15 | 00001101 | 1.1010000 | 3 |
| 17 | 00010001 | 1.0001000 | 4 |
| 19 | 00010011 | 1.0011000 | 4 |
| 138 | 10001010 | 1.0001010 | 7 |
| 63 | 00111111 | 1.1111100 | 5 |

## Postnormalize

## ■ Issue

- Rounding may have caused overflow
- Handle by shifting right once \& incrementing exponent

| Value | Rounded | Exp | Adjusted | Result |
| ---: | :---: | :--- | :--- | :---: |
| 128 | 1.000 | 7 |  | 128 |
| 15 | 1.101 | 3 |  | 15 |
| 17 | 1.000 | 4 |  | 16 |
| 19 | 1.010 | 4 |  | 20 |
| 138 | 1.001 | 7 |  | 134 |
| 63 | 10.000 | 5 | $1.000 / 6$ | 64 |


| Interesting Numbers |  |  | \{single, double\} |
| :---: | :---: | :---: | :---: |
| Description | exp | frac | Numeric Value |
| - Zero | 00... 00 | 00... 00 | 0.0 |
| - Smallest Pos. Denorm. <br> - Single $\approx 1.4 \times 10^{-45}$ <br> - Double $\approx 4.9 \times 10^{-324}$ | 00... 00 | 00... 01 | $2^{-\{23,52\}} \times 2^{-\{126,1022\}}$ |
| - Largest Denormalized <br> - Single $\approx 1.18 \times 10^{-38}$ <br> - Double $\approx 2.2 \times 10^{-308}$ | 00... 00 | 11... 11 | $(1.0-\varepsilon) \times 2^{-\{126,1022\}}$ |
| - Smallest Pos. Normalized <br> - Just larger than largest den | $\begin{aligned} & \text { 00... } 01 \\ & \text { nalized } \end{aligned}$ | 00... 00 | $1.0 \times 2^{-\{126,1022\}}$ |
| - One | 01... 11 | 00... 00 | 1.0 |
| - Largest Normalized <br> - Single $\approx 3.4 \times 10^{38}$ <br> - Double $\approx 1.8 \times 10^{308}$ | 11... 10 | 11... 11 | $(2.0-\varepsilon) \times 2^{\{127,1023\}}$ |

